



*English*

Contact during the exam: Professor Jarle Tufto  
Phone: 99705519

Statistical modelling for biologists and biotechnologists, ST2304

24. mai, 2013

Kl. 9–13

Grades to be announced: 14. juni, 2013

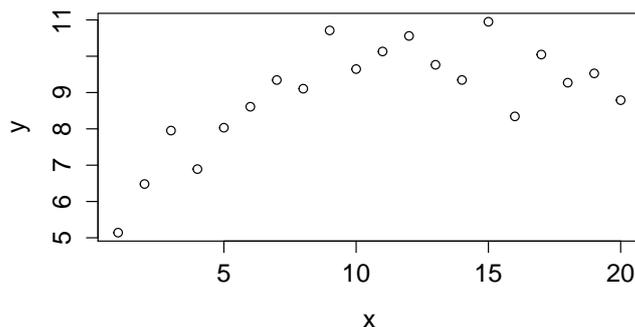
Permitted aids: One handwritten yellow A4 paper, pocket calculator, “Tabeller og formler i statistikk” (Tapir forlag), K. Rottmann: Matematisk formelsamling.

Help pages for some R functions you may need to use follow on page 9.

**Problem 1** We want to test if the variances  $\sigma_X^2$  and  $\sigma_Y^2$  in two normally distributed populations are different from each other and collect two samples of size 10 from the first population and of size 20 from the second. It can be shown that the test statistic  $F = S_X^2/S_Y^2$  where  $S_X^2$  and  $S_Y^2$  are the two sample variances has a  $F$ -distribution with 9 and 19 degrees of freedom under the null hypothesis  $H_0 : \sigma_X^2 = \sigma_Y^2$ .

- a) Write an R expression which computes the two critical values for the test if we choose a level of significance equal to 0.05.
- b) Suppose that the estimate of the two variances are 13.5 and 5.2 respectively. Write an R expression that computes the  $p$ -value of the test.
- c) Does the test statistic have a discrete or continuous distribution? What is the probability that the test statistic takes a on value of exactly 1?
- d) Write an R expression that simulates 1000 realisations of the test statistic under the assumption that  $H_0$  is true and that plots a histogram of these realisations.

**Problem 2** We wish to find the optimal growth temperatur for halibut fry and measure growth  $y$  (grams/week) at 19 different water temperatures  $x$  ( $^{\circ}\text{C}$ ) under otherwise equal conditions. The observed data are shown below.



Suppose that we model the relationship between the dependent variable  $y$  (growth) and temperature by means of multiple regression as follows, using temperature  $x$  and temperature squared  $x^2$  as independent variables (covariates).

```
> x2 <- x^2
> modell <- lm(y ~ x + x2)
> summary(modell)
```

Call:

```
lm(formula = y ~ x + x2)
```

Residuals:

Min	1Q	Median	3Q	Max
-1.55021	-0.31092	0.03203	0.38368	1.04305

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	4.862606	0.497222	9.780	2.14e-08	***
x	0.816051	0.109049	7.483	8.95e-07	***
x2	-0.031351	0.005044	-6.215	9.41e-06	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6683 on 17 degrees of freedom  
 Multiple R-squared: 0.8165, Adjusted R-squared: 0.7949  
 F-statistic: 37.82 on 2 and 17 DF, p-value: 5.504e-07

- a) Write down the model in mathematical (algebraic) notation and state the assumptions being made. Which independent variables (covariates) have a statistically significant effect on the response variable?
- b) Show that the optimal growth temperature  $x_0$  is the function

$$x_0 = f(b_1, b_2) = -\frac{b_1}{2b_2}, \quad (1)$$

where  $b_1$  is the regression coefficient for temperature  $x$  and  $b_2$  is the regression coefficient for temperature squared  $x^2$ . Hint: Equate the derivative of growth  $y$  with respect to temperature  $x$  with zero and solve the resulting equation for  $x$ .

Calculate the estimate  $\hat{x}_0$  of optimal growth temperature. Does the estimate appear reasonable based on the observed data?

- c) Find the standard error of  $\hat{x}_0$ . You will, among other quantities, need an estimate of the covariance between  $\hat{b}_1$  and  $\hat{b}_2$  which can be found from the following R output. See the help page for `vcov` for further information.

```
> vcov(modell)
      (Intercept)      x      x2
(Intercept) 0.247229 -0.048192 0.001959
x           -0.048192 0.011891 -0.000534
x2          0.001959 -0.000534 0.000025
```

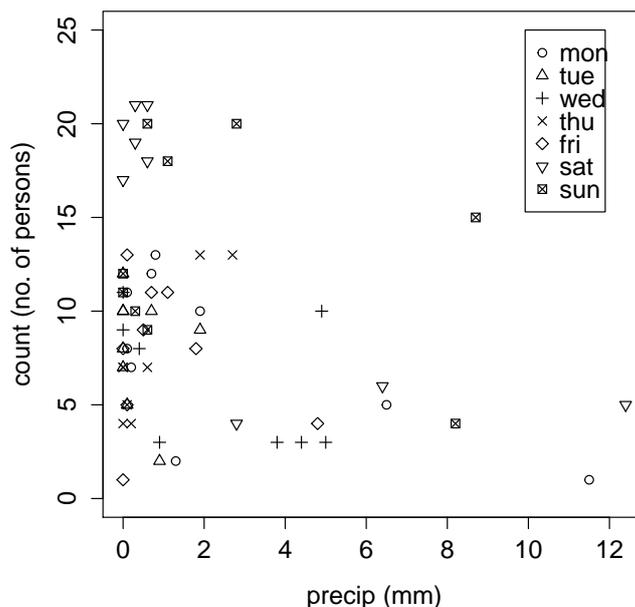
Again, comment on whether the estimate appear reasonable given the above figure.

**Problem 3** As part of a survey looking into recreational use of Bymarka in Trondheim we count the number of walkers, cyclists and runners passing a certain sampling location on their way into Bymarka between 09:00 and 20:00 during each day in June and July 2012. We load the data into the following data frame in R.

```
> bymarka
  weekday weekend precip count
1     fri     no    0.1    13
2     sat     yes    0.3    21
3     sun     yes    2.8    20
4     mon     no    0.2     7
5     tue     no    0.7    10
6     wed     no    0.0    11
7     thu     no    1.9    13
8     fri     no    0.5     9
9     sat     yes    0.0    20
10    sun     yes    0.6    20
11    mon     no    0.1    11
12    tue     no    0.0    10
13    wed     no    0.4     8
14    thu     no    2.7    13
15    fri     no    1.8     8
16    sat     yes    0.6    21
17    sun     yes    8.7    15
18    mon     no    0.8    13
19    tue     no    0.0    10
20    wed     no    0.0     9
21    thu     no    0.0     7
.
.
.
57    fri     no    0.0     1
58    sat     yes   12.4     5
59    sun     yes    8.2     4
60    mon     no    6.5     5
61    tue     no    0.0    12
```

The variables `weekday` and `weekend` are factors representing day of the week and if a day is part of the weekend or not (saturday or sunday) respectively. The variable `precip` is the

amount of precipitation measured in mm and `count` is the number of persons passing the sampling location each day. The data is graphically presented in the following figure.



We first fit the following model (model A).

```
> fitA <- glm(count ~ weekend + precip,fam=poisson)
> summary(fitA)
```

Call:

```
glm(formula = count ~ weekend + precip, family = poisson(link="log"))
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-3.2315	-1.2276	-0.0180	0.9036	2.3345

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	2.12138	0.05893	35.999	< 2e-16 ***
weekendyes	0.69069	0.08493	8.133	4.20e-16 ***
precip	-0.08896	0.01830	-4.860	1.17e-06 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 188.07 on 60 degrees of freedom  
 Residual deviance: 108.11 on 58 degrees of freedom  
 AIC: 352.32

Number of Fisher Scoring iterations: 5

```
> drop1(fitA,test="Chisq")
Single term deletions
```

Model:

```
count ~ weekend + precip
      Df Deviance   AIC   LRT Pr(>Chi)
<none>      108.11 352.32
weekend  1   171.21 413.41 63.093 1.972e-15 ***
precip   1   136.50 378.71 28.390 9.919e-08 ***
```

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

- a) Explain why the Poisson-assumption might be a reasonable starting assumption. How many person are expected to pass the sampling location per day on a weekend compared to on a workday (monday to friday)? Is the difference statistically signifkant? Compute the expected number of persons passing on a day of the weekend given 5mm of precipitation.

To test if there are any additional differences between the expected number of persons passing between the different weekdays beyond the workday/weekend-effect we fit the following alternative model (model B).

```
> fitB <- glm(count ~ weekday + precip,fam=poisson)
> summary(fitB)
```

Call:

```
glm(formula = count ~ weekday + precip, family = poisson)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-3.2586	-1.1416	-0.0993	0.9164	2.4785

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	2.22000	0.12422	17.871	< 2e-16 ***
weekdaytue	-0.09275	0.16958	-0.547	0.584407
weekdaywed	-0.19018	0.18585	-1.023	0.306167
weekdaythu	-0.14767	0.17758	-0.832	0.405633
weekdayfri	-0.08671	0.17030	-0.509	0.610648
weekdaysat	0.63720	0.14876	4.283	1.84e-05 ***
weekdaysun	0.54393	0.15132	3.595	0.000325 ***
precip	-0.08874	0.01857	-4.777	1.78e-06 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 188.07 on 60 degrees of freedom  
 Residual deviance: 106.33 on 53 degrees of freedom  
 AIC: 360.54

Number of Fisher Scoring iterations: 5

&gt; drop1(fitB,test="Chisq")

Single term deletions

Model:

	Df	Deviance	AIC	LRT	Pr(>Chi)
<none>		106.33	360.54		
weekday	6	171.21	413.41	64.877	4.572e-12 ***
precip	1	133.60	385.81	27.269	1.770e-07 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

- b) Explain why model A is nested in model B. What are the number of estimated parameters under model A and B. Test if there is any difference between weekdays or between saturday/sundays in the expected number of persons passing beyond the workday/weekend-effect. Hint: The result of this test can not be read directly from the above output but must be computed "by hand".

- c) Test if there is overdispersion in the data under the selected model. Discuss possible mechanisms that might generate overdispersion in this specific case.

FDist package:stats R Documentation

The F Distribution

Description:

Density, distribution function, quantile function and random generation for the F distribution with 'df1' and 'df2' degrees of freedom (and optional non-centrality parameter 'ncp').

Usage:

```
df(x, df1, df2, ncp, log = FALSE)
pf(q, df1, df2, ncp, lower.tail = TRUE, log.p = FALSE)
qf(p, df1, df2, ncp, lower.tail = TRUE, log.p = FALSE)
rf(n, df1, df2, ncp)
```

Arguments:

x, q: vector of quantiles.

p: vector of probabilities.

n: number of observations. If 'length(n) > 1', the length is taken to be the number required.

df1, df2: degrees of freedom. 'Inf' is allowed.

ncp: non-centrality parameter. If omitted the central F is assumed.

log, log.p: logical; if TRUE, probabilities p are given as log(p).

lower.tail: logical; if TRUE (default), probabilities are P[X <= x], otherwise, P[X > x].

Details:

The F distribution with 'df1 = ' n1 and 'df2 = ' n2 degrees of freedom has density

$$f(x) = \frac{\Gamma((n1 + n2)/2)}{\Gamma(n1/2) \Gamma(n2/2)} (n1/n2)^{n1/2} x^{n1/2 - 1} (1 + (n1/n2) x)^{-(n1 + n2)/2}$$

for  $x > 0$ .

It is the distribution of the ratio of the mean squares of n1 and n2 independent standard normals, and hence of the ratio of two independent chi-squared variates each divided by its degrees of freedom. Since the ratio of a normal and the root mean-square of m independent normals has a Student's  $t_m$  distribution, the square of a  $t_m$  variate has a F distribution on 1 and m degrees of freedom.

The non-central F distribution is again the ratio of mean squares of independent normals of unit variance, but those in the numerator are allowed to have non-zero means and 'ncp' is the sum of squares of the means. See Chisquare for further details on non-central distributions.

Value:

'df' gives the density, 'pf' gives the distribution function 'qf' gives the quantile function, and 'rf' generates random deviates.

Invalid arguments will result in return value 'NaN', with a warning.

Note:

Supplying 'ncp = 0' uses the algorithm for the non-central distribution, which is not the same algorithm used if 'ncp' is omitted. This is to give consistent behaviour in extreme cases with values of 'ncp' very near zero.

The code for non-zero 'ncp' is principally intended to be used for moderate values of 'ncp': it will not be highly accurate, especially in the tails, for large values.

Source:

For the central case of 'df', computed via a binomial

probability, code contributed by Catherine Loader (see 'dbinom'); for the non-central case computed via 'dbeta', code contributed by Peter Ruckdeschel.

For 'pf', via 'pbeta' (or for large 'df2', via 'pchisq').

For 'qf', via 'qchisq' for large 'df2', else via 'qbeta'.

References:

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988) *The New S Language*. Wadsworth & Brooks/Cole.

Johnson, N. L., Kotz, S. and Balakrishnan, N. (1995) *Continuous Univariate Distributions*, volume 2, chapters 27 and 30. Wiley, New York.

See Also:

Distributions for other standard distributions, including 'dchisq' for chi-squared and 'dt' for Student's t distributions.

Examples:

```
## the density of the square of a t_m is 2*dt(x, m)/(2*x)
# check this is the same as the density of F_{1,m}
x <- seq(0.001, 5, len=100)
all.equal(df(x^2, 1, 5), dt(x, 5)/x)

## Identity: qf(2*p - 1, 1, df) == qt(p, df)^2 for p >= 1/2
p <- seq(1/2, .99, length=50); df <- 10
rel.err <- function(x,y) ifelse(x==y,0, abs(x-y)/mean(abs(c(x,y))))
quantile(rel.err(qf(2*p - 1, df1=1, df2=df), qt(p, df)^2), .90) # ~ 7e-9
```

----- package:stats R Documentation

Calculate Variance-Covariance Matrix for a Fitted Model Object

Description:

Returns the variance-covariance matrix of the main parameters of a fitted model object.

Usage:

```
vcov(object, ...)
```

Arguments:

object: a fitted model object, typically. Sometimes also a 'summary()' object of such a fitted model.

...: additional arguments for method functions. For the 'glm' method this can be used to pass a 'dispersion' parameter.

Details:

This is a generic function. Functions with names beginning in 'vcov.' will be methods for this function. Classes with methods for this function include: 'lm', 'mlm', 'glm', 'nls', 'summary.lm', 'summary.glm', 'negbin', 'polr', 'rlm' (in package 'MASS'), 'multinom' (in package 'nnet') 'gls', 'lme' (in package 'nlme'), 'coxph' and 'survreg' (in package 'survival').

('vcov()') methods for summary objects allow more efficient and still encapsulated access when both 'summary(mod)' and 'vcov(mod)' are needed.)

Value:

A matrix of the estimated covariances between the parameter estimates in the linear or non-linear predictor of the model. This should have row and column names corresponding to the parameter names given by the 'coef' method.